

EE5110 Segment 3

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Q1

a) Calculus of variations

The input signal is :

$$u = -\frac{a}{2}y + \frac{\dot{y}}{2}$$

The function $J(y, u)$ changes to:

$$\begin{aligned} J(y) &= \int_0^\infty [y^2 + (-\frac{a}{2}y + \frac{\dot{y}}{2})^2] dt \\ &= \int_0^\infty [(1 + \frac{a^2}{4})y^2 + \frac{1}{4}\dot{y}^2] dt - \frac{a}{2} \int_0^\infty y\dot{y} dt \end{aligned}$$

\therefore

$$\int_0^\infty y\dot{y} dt = y^2|_0^\infty - \int_0^\infty y\dot{y} dt$$

$$\int_0^\infty y\dot{y} dt = \frac{1}{2}[y^2(\infty) - y^2(0)]$$

\therefore

$$J(y) = \int_0^\infty [(\frac{a^2}{4} + 1)y^2 + \frac{1}{4}\dot{y}^2] dt - \frac{a}{4}y^2(\infty) + \frac{a}{4}c^2$$

Let $z(t)$ denote any function of t with the property that $J(z)$ exists. Take ε to be a scalar parameter.

$$J(y_0 + \varepsilon z) = \int_0^\infty [(\frac{a^2}{4} + 1)(y_0 + \varepsilon z)^2 + \frac{1}{4}(\dot{y}_0 + \varepsilon \dot{z})^2] dt - \frac{a}{4}y^2(\infty) + \frac{a}{4}c^2$$

The $J(y_0 + \varepsilon z)$ must have an absolute minimum at $\varepsilon = 0$

$$\frac{d}{d\varepsilon} J(y_0 + \varepsilon z)|_{\varepsilon=0} = 0$$

$$\begin{aligned} J(y_0 + \varepsilon z) &= \int_0^\infty [(\frac{a^2}{4} + 1)y_0^2 + \frac{1}{4}\dot{y}_0^2] dt + 2\varepsilon \int_0^\infty [(\frac{a^2}{4} + 1)y_0 z + \frac{a^2}{4}\dot{y}_0 \dot{z}] dt \\ &\quad + \varepsilon^2 \int_0^\infty [(\frac{a^2}{4} + 1)z^2 + \frac{a^2}{4}\dot{z}^2] dt - \frac{a^2}{4}y^2(\infty) + \frac{a^2}{4}c^2 \end{aligned}$$

We see then that the variational condition derived is :

$$\int_0^\infty [(\frac{a^2}{4} + 1)y_0^2 + \frac{1}{4}\dot{y}_0^2] dt = 0$$

$$2\varepsilon \int_0^\infty [(\frac{a^2}{4} + 1)y_0 z + \frac{a^2}{4}\dot{y}_0 \dot{z}] dt = 0$$

$$\int_0^\infty [(\frac{a^2}{4} + 1)z^2 + \frac{a^2}{4}\dot{z}^2] dt - \frac{a^2}{4}y^2(\infty) + \frac{a^2}{4}c^2 = 0$$

Since $y_0 + \varepsilon z$ is an admissible function satisfies the initial condition :

$$y_0(0) + \varepsilon z(0) = c$$

We see that $z(0) = 0$.

Since the left - hand side must be zero for all admissible z , we suspect that

$$(\frac{a^2}{4} + 1)y_0 - \frac{1}{4}\ddot{y}_0 = 0$$

First, we use T to replace ∞ , $\dot{y}_0(T) = 0$. And we obtain no condition on $\dot{y}_0(0)$. We can get :

$$y_0(0) = c, \quad \dot{y}_0(T) = 0$$

The general solution of the differential equation is :

$$y = c_1 e^{\sqrt{4+a^2} \cdot t} + c_2 e^{-\sqrt{4+a^2} \cdot t}$$

Using the boundary conditions, we have the two equations to determine the coefficients c_1 and c_2 .

$$\begin{aligned} c &= c_1 + c_2 \\ 0 &= c_1 e^{\sqrt{4+a^2} \cdot T} - c_2 e^{-\sqrt{4+a^2} \cdot T} \end{aligned}$$

Solving, we obtain the expression :

$$y_o(t) = c \left(\frac{e^{\sqrt{4+a^2}(t-T)} + e^{-\sqrt{4+a^2}(t-T)}}{e^{\sqrt{4+a^2} \cdot T} + e^{-\sqrt{4+a^2} \cdot T}} \right) = c \frac{\cosh(\sqrt{4+a^2}(t-T))}{\cosh(\sqrt{4+a^2} \cdot T)}$$

Let $T \rightarrow \infty$, We have

$$\begin{aligned} y_o(t) &= c \left(\frac{e^{\sqrt{4+a^2}(t-T)} + e^{-\sqrt{4+a^2}(t-T)}}{e^{\sqrt{4+a^2} \cdot T} + e^{-\sqrt{4+a^2} \cdot T}} \right) = c \left(\frac{e^{\sqrt{4+a^2}(t-2T)} + e^{-\sqrt{4+a^2} \cdot t}}{e^{-2\sqrt{4+a^2} \cdot T} + 1} \right) \rightarrow c e^{-\sqrt{4+a^2} \cdot t} \\ \dot{y}_o(t) &= c \sqrt{4+a^2} \cdot \left(\frac{e^{\sqrt{4+a^2}(t-T)} - e^{-\sqrt{4+a^2}(t-T)}}{e^{\sqrt{4+a^2} \cdot T} + e^{-\sqrt{4+a^2} \cdot T}} \right) = c \sqrt{4+a^2} \cdot \left(\frac{e^{\sqrt{4+a^2}(t-2T)} - e^{-\sqrt{4+a^2} \cdot t}}{e^{-2\sqrt{4+a^2} \cdot T} + 1} \right) \rightarrow -c \sqrt{4+a^2} \cdot e^{-\sqrt{4+a^2} \cdot t} \\ \dot{y}_o &= -\sqrt{4+a^2} \cdot y_o(t) \end{aligned}$$

So we have the control laws

b) Dynamic programming

Optimal Value function :

$$V(c, T) = \min_y J(y)$$

$$J(y) = \int_0^\Delta + \int_\Delta^T = (c^2 + u^2)\Delta + V(c + (ac + 2u)\Delta, T - \Delta) + O(\Delta^2)$$

We can use Taylor series to relate $V(c + (ac + 2u)\Delta, T - \Delta)$ with $V(c, T)$, $J(y)$ will change to

$$V(c, T) = \min_u [(c^2 + u^2)\Delta + V(c, T) + \frac{\partial V}{\partial c}(ac + 2u)\Delta - \frac{\partial V}{\partial T}\Delta + O(\Delta^2)]$$

Ignoring the higher order terms of Δ , we have

$$\frac{\partial V}{\partial T} = \min_u [(c^2 + u^2) + \frac{\partial V}{\partial c}(ac + 2u)]$$

When $T \rightarrow \infty$, $V(c, T)$ becomes $V(c)$,

$$\begin{aligned} V(c) &= \min_u [(c^2 + u^2)\Delta + V(c + (ac + 2u)\Delta)] + O(\Delta^2) \\ 0 &= \min_u [(c^2 + u^2) + \dot{V}(c)(ac + 2u)] \end{aligned}$$

Take the derivative respect to u gives $2u + 2\dot{V}(c)a = 0$, so

$$\begin{aligned} u &= -\dot{V}(c) \\ 0 &= (c^2 + (\dot{V}(c))^2) + \dot{V}(c)(ac - 2\dot{V}(c)) \\ \dot{V}^2(c) - ac\dot{V}(c) + c^2 &= 0 \end{aligned}$$

So we have two possibilities, with the condition $V(0) = 0$, we can obtain two possible solutions :

$$\begin{aligned} V(c) &= \frac{a + \sqrt{a^2 + 4}}{2} \\ V(c) &= \frac{a - \sqrt{a^2 + 4}}{2} \end{aligned}$$

Since $V(c) \geq 0$, we see that $V(c) = \frac{a + \sqrt{a^2 + 4}}{2}$, the optimal value can be easily obtained as

$$u = -\frac{a + \sqrt{a^2 + 4}}{2}c$$

Since $y(0) = c$, so we have $u(0) = \frac{a + \sqrt{a^2 + 4}}{2} y(0)$. At any time t , we will have the control law :

$$u(t) = -\frac{a + \sqrt{a^2 + 4}}{2} y(t)$$

The results from two method are same

Q2

We write the optimal value function as

$$V_N(c) = \min_{u_n} J_N(y, u)$$

After $u(0)$ is chosen, the new state of the system is $y(1) = 2c + au(0)$, The cost function takes the form

$$c^2 + u^2(0) + \sum_{n=1}^N (y^2(n) + u^2(n))$$

The long term cost can be expressed as optimal value starting from $2c + au(0)$ with $N - 1$ steps left

$$\sum_{n=1}^N (y^2(n) + u^2(n)) = V_{N-1}(ac + 2u(0))$$

Then

$$V_N(c) = \min_{u(0)} [c^2 + u^2(0) + V_{N-1}(ac + 2u(0))]$$

For the continuous case we have $V(c, T) = c^2 r(T)$

It is reasonable to guess that

$$V_N(c) = c^2 r_N$$

$$c^2 r_N = \min_{u(0)} [c^2 + u^2(0) + (ac + 2u(0))^2 r_{N-1}]$$

The value of $u(0)$ that minimizes is readily obtained by differentiation

$$2u(0) + 2a(2a + au(0))r_{N-1} = 0$$

$$u(0) = -\frac{2acr_{N-1}}{1 + 4r_{N-1}}$$

Using this value, we obtain the recurrence relation

$$r_N = 1 + \frac{a^2 r_{N-1}}{1 + 4r_{N-1}}$$

At each time $t = k$, the input control is

$$u(k) = -\frac{2acr_{N-k-1}y(k)}{1 + 4r_{N-k-1}}$$

When $N \rightarrow \infty$, let $r = \lim_{N \rightarrow \infty} r_N$, then r is the positive root of the quadratic equation

$$r = \frac{(a^2 + 3) + \sqrt{a^4 + 6a^2 + 25}}{8}$$

The control signal will change to :

$$\lim_{N \rightarrow \infty} u(0) = -\frac{2acr}{1 + 4r}$$

\therefore for the infinite time process, the optimal feedback controller is :

Q3

Assume that the lifeguard will run to $(a, 0)$, and then swim to the swimmer. The parameter we can get from question : v /s. The optimal function can be expressed as :

$$T = \min_a \left[\left(\frac{\sqrt{a^2 + 10^2}}{v_1} \right) + \left(\frac{\sqrt{(20-a)^2 + (-10)^2}}{v_2} \right) \right]$$

Take the derivative respect to a , we get

$$\frac{29a^2}{100} - 10a + 129 = 0$$

$$a = \frac{10}{0.58} = 17.24$$

So, the shortest time path is that lifeguard run to $(18.823, 0)$ and then swim to the swimmer. The shortest time is :

$$t_{min} = \frac{\sqrt{a^2 + 10^2}}{v_1} + \frac{\sqrt{(20-a)^2 + (-20)^2}}{v_2} = 9.17s$$



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In [1]: import numpy as np
np.sqrt(17.24**2+100)/5+np.sqrt((20-17.24)**2+100)/2
```

Out[1]: 9.173008873571147

Q4

Method1: Programming view

This question we can refer to Dijkstra, Floyd method. It has the same principle with optimal solution method.

The basic logic is :

Based on the algorithm, suppose we get the optimal result of set S , S includes the places we traveled.

We have n place, suppose we traveled k places, $k < n$ and belongs to S , the $S(k)$ should also be optimal result.

We can put all attractions and hotel in the x y plane and sort them by x coordinate from small to large $p_0, p_1, p_2, \dots, p_n$.

In order to get the shortest circle, we set $\text{dist}(i, j) = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$ as the distance between place i and j .

step1 : Set S contains only the source point. $S = p_0$, and the distance of p_0 is 0. U contains vertices other than places, $i.e.$ and we can calculate all the distance between two places.

step2

: Select a vertex p_k from U with the smallest distance from p_0 and add p_k to S (the selected distance is the shortest path

step3 : Take k as the newly considered intermediate point and modify the distance of each vertex in U ;

If the distance from the source point p_0 to the vertex p_u (through the vertex p_k) is shorter than the original distance (without passing through the vertex p_k), the distance value of the vertex p_u is modified,

and the modified distance value is the distance of the vertex p_k plus the distance between two places.

step4 : Repeat the step2 and step 3 until all the places are in S set.

Method2: Mathematic view

First, we can put all attractions and hotel in the x y plane and sort them by x coordinate from small to large p_0, p_1, p_2, \dots

Assume that $V_{i,j}$ ($i \leq j$) is the shortest closed curve which contain p_0, p_1, \dots, p_n .

This path goes from p_i to p_0 , and then goes from p_0 right to p_j . So, $V_{n,n}$ is what we want in this topic.

Assume that the length of $V_{i,j}$ is $l(i, j)$, the distance between p_i and p_j is $\text{dist}(i, j) = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$

In the path $V_{i,j}$, p_i is in the path $p_i \rightarrow p_1$, p_j is in the path $p_1 \rightarrow p_j$. Now, let's talk about the position of p_{j-1}

(1) $i < j - 1$

Because p_{j-1} is on the right side of p_i , so p_{j-1} is in the path $p_1 \rightarrow p_j$.

Besides, p_{j-1} is the rightmost point except p_j , so it connect to p_j directly. We can get

$$l(i, j) = l(i, j - 1) + \text{dist}(j - 1, j)$$

(2) $i = j - 1$

In this case, p_{j-1} is p_i , so p_{j-1} is in the path $p_i \rightarrow p_1$. Any point from p_1, p_2, \dots, p_{j-2} can connect to $p + j$.

Assume that point is p_k ($1 \leq k \leq j - 2$). We need to chose an appropriate point p_k so that we can get the shortest $l(i, j)$

$$l(i, j) = \min_{1 \leq k \leq j-2} [l(k, j - 1) + \text{dist}(k, j)]$$

(3) $i = j$

This only happens when $i = j = n$. In this case, p_{n-1} connect to p_n , we can get :

$$l(n, n) = l(n - 1, n) + \text{dist}(n - 1, n)$$

In conclusion the optimal function is :

$$l(i, j) = \begin{cases} l(i, j - 1) + \text{dist}(j - 1, j), & i < j - 1 \\ \min_{1 \leq k \leq j-2} [l(k, j - 1) + \text{dist}(k, j)], & i = j - 1 \\ l(n - 1, n) + \text{dist}(n - 1, n), & i = j = n \end{cases}$$

This function is what we want.

